

# **Application of Information Theory, Lecture 11**

## **Pseudo-Entropy and Pseudorandom Generators**

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# Part I

## Motivation

# Encryption schemes

## Definition 1

A pair of algorithms  $(E, D)$  is (perfectly correct) encryption scheme, if for any  $k \in \{0, 1\}^n$  and  $m \in \{0, 1\}^\ell$ , it holds that  $D(k, E(k, m)) = m$

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- ▶ Statistical security? HW. Computational security?

## Part II

# Statistical Vs. Computational distance

## Distributions and statistical distance

Let  $\mathcal{P}$  and  $\mathcal{Q}$  be two distributions over a finite set  $\mathcal{U}$ . Their **statistical distance** (also known as, variation distance) is defined as

$$\text{SD}(\mathcal{P}, \mathcal{Q}) := \frac{1}{2} \sum_{x \in \mathcal{U}} |\mathcal{P}(x) - \mathcal{Q}(x)| = \max_{S \subseteq \mathcal{U}} (\mathcal{P}(S) - \mathcal{Q}(S))$$

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### Claim 2

For any pair of (finite) distribution  $\mathcal{P}$  and  $\mathcal{Q}$ , it holds that

$$\text{SD}(\mathcal{P}, \mathcal{Q}) = \max_D \{ \Delta^D(\mathcal{P}, \mathcal{Q}) := \Pr_{x \leftarrow \mathcal{P}} [D(x) = 1] - \Pr_{x \leftarrow \mathcal{Q}} [D(x) = 1] \},$$

where  $D$  is **any** algorithm.



## Some useful facts

Let  $\mathcal{P}, \mathcal{Q}, \mathcal{R}$  be finite distributions, then

**Triangle inequality:**  $\text{SD}(\mathcal{P}, \mathcal{R}) \leq \text{SD}(\mathcal{P}, \mathcal{Q}) + \text{SD}(\mathcal{Q}, \mathcal{R})$

**Repeated sampling:**  $\text{SD}(\mathcal{P}^2 = (\mathcal{P}, \mathcal{P}), \mathcal{Q}^2 = (\mathcal{Q}, \mathcal{Q})) \leq 2 \cdot \text{SD}(\mathcal{P}, \mathcal{Q})$

# Section 1

## **Computational Indistinguishability**

# Computational indistinguishability

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- ▶ Can it be different from the statistical case?
- ▶ Unless said otherwise, distributions are over  $\{0, 1\}^n$



## Repeated sampling

### Question 4

Assume  $\mathcal{P}$  and  $\mathcal{Q}$  are  $(s, \epsilon)$ -indistinguishable, what about  $\mathcal{P}^2$  and  $\mathcal{Q}^2$ ?

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- ▶ So either  $\Delta_{(\mathcal{P}^2, (\mathcal{P}, \mathcal{Q}))}^D \geq \varepsilon'/2$ , or  $\Delta_{((\mathcal{P}, \mathcal{Q}), \mathcal{Q}^2)}^D \geq \varepsilon'/2$

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$$\begin{aligned}\varepsilon' &= \Pr_{x \leftarrow \mathcal{P}^2} [D(x) = 1] - \Pr_{x \leftarrow \mathcal{Q}^2} [D(x) = 1] \\ &= \left( \Pr_{x \leftarrow \mathcal{P}^2} [D(x) = 1] - \Pr_{x \leftarrow (\mathcal{P}, \mathcal{Q})} [D(x) = 1] \right) \\ &\quad + \left( \Pr_{x \leftarrow (\mathcal{P}, \mathcal{Q})} [D(x) = 1] - \Pr_{x \leftarrow \mathcal{Q}^2} [D(x) = 1] \right) \\ &= \Delta_{(\mathcal{P}^2, (\mathcal{P}, \mathcal{Q}))}^D + \Delta_{((\mathcal{P}, \mathcal{Q}), \mathcal{Q}^2)}^D\end{aligned}$$

- ▶ So either  $\Delta_{(\mathcal{P}^2, (\mathcal{P}, \mathcal{Q}))}^D \geq \varepsilon'/2$ , or  $\Delta_{((\mathcal{P}, \mathcal{Q}), \mathcal{Q}^2)}^D \geq \varepsilon'/2$
- ▶ Hence,  $\varepsilon' < 2\varepsilon$  implies that  $s' \geq s - 2n$ .

# Repeated sampling

## Question 4

Assume  $\mathcal{P}$  and  $\mathcal{Q}$  are  $(s, \varepsilon)$ -indistinguishable, what about  $\mathcal{P}^2$  and  $\mathcal{Q}^2$ ?

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- ▶ More generally,  $\mathcal{P}^k$  and  $\mathcal{Q}^k$  are  $(s - nk, k\varepsilon)$ -indistinguishable.
- ▶ In the uniform settings things behaves very differently!

# Part III

## Pseudorandom Generators

# Pseudorandom generator

## Definition 5 (pseudorandom distributions)

A distribution  $\mathcal{P}$  over  $\{0, 1\}^n$  is  $(s, \epsilon)$ -pseudorandom, if it is  $(s, \epsilon)$ -indistinguishable from  $U_n$ .

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A poly-time computable function  $g: \{0, 1\}^n \mapsto \{0, 1\}^{\ell(n)}$  is a  $(s, \varepsilon)$ -pseudorandom generator, if for any  $n \in \mathbb{N}$

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- ▶ Do such generators exist?
- ▶ Applications?

## Section 2

# **Pseudorandom generators (PRGs) from One-Way Permutations (OWPs)**

## OWP to PRG

### Claim 7

Let  $f: \{0, 1\}^n \mapsto \{0, 1\}^n$  be a poly-time permutation and let  $b: \{0, 1\}^n \mapsto \{0, 1\}$  be a poly-time  $(s, \varepsilon)$ -hardcore predicate of  $f$ , then  $g(x) = (f(x), b(x))$  is a  $(s - O(n), \varepsilon)$ -PRG.

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- ▶ Proof: Let  $D$  be an  $s'$ -size algorithm with  $\Delta_{g(U_n), U_{n+1}}^D = \varepsilon'$ , we will show  $\exists (s' + O(n))$ -size  $P$  with  $\Pr[P(f(U_n)) = b(U_n)] = \frac{1}{2} + \varepsilon'$ .



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$$\delta = \Pr[D(f(U_n), U_1) = 1]$$

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### Algorithm 8 (P)

Input:  $y \in \{0, 1\}^n$

1. Flip a random coin  $c \leftarrow \{0, 1\}$ .
2. If  $D(y, c) = 1$  output  $c$ , otherwise, output  $\bar{c}$ .



## OWP to PRG cont.

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# Part IV

## **PRG from Regular OWF**

# Computational notions of entropy

## Definition 9

$X$  has  $(s, \varepsilon)$ -pseudoentropy at least  $k$ , if  $\exists$  rv  $Y$  with  $H(Y) \geq k$  and  $\Delta^D(X, Y) \leq \varepsilon$  for any  $s$ -size  $D$ .

$(s, \varepsilon)$ -pseudo min/Reiny -entropy are analogously defined.

# Computational notions of entropy

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### ► Examples

# Computational notions of entropy

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- ▶ Examples
- ▶ Repeated sampling
- ▶ Ensembles
- ▶ In the following we will simply write  $(s, \epsilon)$ -entropy, etc



# High entropy OWF from regular OWF

## Claim 10

Let  $f: \{0, 1\}^n \mapsto \{0, 1\}^n$  be a  $2^k$ -regular  $(s, \varepsilon)$ -one-way, let  $\mathcal{H} = \{h: \{0, 1\}^n \mapsto \{0, 1\}^{m=k+\lceil \log n \rceil}\}$  be 2-universal family, and let  $g(h, x) = (g(x), h, h(x))$ . Then

1.  $H_2(g(U_n, H)) \geq 2n - \frac{1}{n}$ , for  $H \leftarrow \mathcal{H}$ .
2.  $g$  is  $(\Theta(s\varepsilon^2/n), 2\varepsilon)$ -one-way.

- ▶  $k$  and  $m$  and  $\mathcal{H}$  are parameterized by of  $n$
- ▶ We assume  $\log |\mathcal{H}| = n$  and  $s \geq n$

$g$  has high Renyi entropy

## $g$ has high Renyi entropy

$$\begin{aligned}\text{CP}(g(U_n, H)) &:= \Pr_{w, w' \leftarrow \{0,1\}^n \times \mathcal{H}} [g(w) = g(w')] \\ &= \Pr_{h, h' \leftarrow \mathcal{H}} [h = h'] \cdot \Pr_{(x, x') \leftarrow (\{0,1\}^n)^2} [f(x) = f(x')] \\ &\quad \cdot \Pr_{h \leftarrow \mathcal{H}; (x, x') \leftarrow (\{0,1\}^n)^2} [h(x) = h(x') \mid f(x) = f(x')]\end{aligned}$$

## $g$ has high Renyi entropy

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Hence,  $H_2(g(U_n, H)) \geq H_2(\mathcal{H}) + H_2(U_n) + \log \frac{1}{1 + \frac{1}{n}} \geq n + n - \frac{1}{n}$ .

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Consider the following inverter for  $f$

### Algorithm 11 (B)

Input:  $y \in \{0, 1\}^n$ .

Return  $D(y, h, z)$ , for  $h \leftarrow \mathcal{H}$  and  $z \leftarrow \{0, 1\}^\ell$ .

### Algorithm 12 (D)

Input:  $y \in \{0, 1\}^n$ ,  $h \in \mathcal{H}$  and  $z_1 \in \{0, 1\}^\ell$ .

For all  $z_2 \in \{0, 1\}^{m-\ell}$ :

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- ▶  $\Pr_{x \leftarrow \{0,1\}^n; h \leftarrow \mathcal{H}} [D(f(x), h, h(x)_{1,\dots,\ell}) \in f^{-1}(f(x))] \geq \varepsilon'$

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We saw that

$$\Pr_{x \leftarrow \{0,1\}^n; h \leftarrow \mathcal{H}} [D(f(x), h, h(x)_1, \dots, \ell) \in f^{-1}(f(x))] \geq \epsilon' \quad (1)$$

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$$\text{SD}((f(x), h, h(x)_1, \dots, h(x)_\ell)_{x \leftarrow \{0,1\}, h \leftarrow \mathcal{H}}, (f(x), h, U_\ell)_{x \leftarrow \{0,1\}, h \leftarrow \mathcal{H}}) \leq \varepsilon'/2 \quad (2)$$

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Hence,

$$\Pr_{x \leftarrow \{0,1\}^n} [B(f(x)) \in f^{-1}(f(x))] \geq \varepsilon' - \varepsilon'/2 = \varepsilon'/2.$$



# The generator

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Let  $g: \{0, 1\}^n \mapsto \{0, 1\}^m$  be a function with  $H_2(f(U_n)) \geq n - \frac{1}{2}$ , and let  $b$  be  $(s, \epsilon)$ -hardcore predicate for  $g$ . Then  $v(U_n) = (g(U_n), b(U_n))$  has  $(s, \epsilon)$ -Renyi-entropy  $n + \frac{1}{2}$ .

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If  $f$  and  $b$  and  $\mathcal{H}$  (?) are poly-time computable, then  $G$  is a  $(s - ns_v - s_{\mathcal{H}}, n\varepsilon + 2^{-n/4})$ -PRG.

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- ▶ By the leftover hash lemma  $\text{SD}((H, H(Z^n)), (H, U_{n^2+n/4})) \leq 2^{-n/4}$
- ▶ Let  $D$  be an  $s'$ -size algorithm that distinguishes  $G(U_n^n, H)$  from  $(H, U_{n^2+n/4})$  with advantage  $\varepsilon' + 2^{-n/4}$



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- ▶ Hence,  $\exists (s' + s_{\mathcal{H}})$ -size algorithm that distinguishes  $G(U_n^n, H)$  from  $(H, H(Z^n))$  with advantage  $\varepsilon'$

## The generator cont.

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Let  $\mathcal{H}: \{0, 1\}^{n^2+n} \mapsto \{0, 1\}^{n^2+n/4}$  be an 2-universal family and let  $G: \{0, 1\}^n \times \mathcal{H}$  defined by  $G(x_1, \dots, x_n, h) = (h, h(v^n(x_1, \dots, x_n)))$ . Then  $G(H, U_n^n)$  is  $(s - ns_v - s_{\mathcal{H}}, n\varepsilon + 2^{-n/4})$  indistinguishable from  $(H, U_{n^2+n/4})$ , for  $H \leftarrow \mathcal{H}$  and  $s_{\mathcal{H}}$  being the sampling and evaluating time of  $\mathcal{H}$ .

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Proof:

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